

Solution 1

Assume first that we still make the simplifying assumption about queries in that no query point shares an x -coordinate with a segment or is on a segment. Removing the general position assumption allows for vertical line segments and segments intersecting in their endpoints. We have to change the definition of the trapezoidal decomposition slightly: For every endpoint $p \in P(S)$ that is not an endpoint of some vertical segment we draw two vertical extensions from p , one up and one down, that end when meeting the first element of $P(S) \cup S^\circ$ or that extend to infinity if no element is met. If $p \in P(S)$ is the lower (upper) endpoint of some vertical segment, and p is not the upper (lower) endpoint any vertical segment, we draw only one vertical extension from p going downwards (upwards). As in the lecture notes, the maximal open connected regions resulting from removing the segments and the extensions form the trapezoidal decomposition. We call these regions trapezoids even though they could be for example triangles.

With the new definition all the nice properties that we have proved still hold but we can remove the general position assumption. Lemma 3.8. Observation 3.9. and Observation 3.10. are easily seen to hold. All in all we get Theorem 3.11. also without the general position assumption.

Let us finally consider the case when we do not make the simplifying assumption about the query points. In that case we need to decide for every point on the boundary of some trapezoids, to which trapezoid the point belongs to. Informally, we extend some of the trapezoids, the maximal open regions, to also contain parts of their boundary and how this should be done depends on the goal. We show one way to do it.

We let every point in the relative interior of a non-vertical segment belong to the trapezoid below the segment. We also let every point in the relative interior of a vertical extension or the relative interior of a vertical segment belong to the trapezoid that is immediately to its right. Lastly, we let every endpoint of a segment belong to the same trapezoid that the vertical extension immediately below it or the vertical segment below it belongs to. This gives a partition of the whole plane \mathbf{R}^2 into $O(n)$ many connected regions. The same analysis as before goes through with minor changes.

Solution 2

- (a) Suppose $R(\mathcal{S}) \neq \emptyset$. Clearly, this implies that $R(\mathcal{S} \setminus \{H\}) \neq \emptyset$. Therefore and since $H_1, H_2 \in \mathcal{S} \setminus \{H\}$, both $x := c(\mathcal{S})$ and $x' := c(\mathcal{S} \setminus \{H\})$ are well-defined points. Now there are two cases. Either $x' \in H$ or $x' \notin H$.

Suppose $x' \in H$. In that case, we readily have $x' \in R(\mathcal{S})$. And since x' was the point of smallest x -coordinate in $R(\mathcal{S} \setminus \{H\}) \supseteq R(\mathcal{S})$, it will a fortiori be the point of smallest x -coordinate in $R(\mathcal{S})$. Therefore $x' = x$ in that case.

Now suppose $x' \notin H$. In that case, we have $x' \notin R(\mathcal{S})$. Now consider the line segment s connecting x' and x . Since $x \in H$ but $x' \notin H$, s must cross ℓ_H at an intersection point p . Moreover, since $x', x \in R(\mathcal{S} \setminus \{H\})$ and since $R(\mathcal{S} \setminus \{H\})$ is a convex set, the whole line segment s runs inside $R(\mathcal{S} \setminus \{H\})$. Since $p \in H$ by construction, it follows $p \in R(\mathcal{S})$. Clearly, x' must have a strictly smaller x -coordinate than x , otherwise x would have been a possible choice for $c(\mathcal{S} \setminus \{H\})$ in the first place. Therefore, going further away from x' along s than p would only make x -coordinates larger, from which we conclude $p = x$ and thus $x \in \ell_H$.

- (b) For any halfplane $G \in \mathcal{S} \setminus \{H\}$, let us say that G *points right* (with respect to H) if the intersection of G and ℓ_H is a ray that extends infinitely to the right. Similarly, let us say that G *points left* (with respect to H) if the intersection of G and ℓ_H is a ray that extends infinitely to the left. There is no third option because we assumed not to have vertical or parallel lines.

We can now proceed in the following way:

SolveInc. Input: $\mathcal{S}, H, x' = c(\mathcal{S} \setminus \{H\})$.

- (1) Check whether $x' \in H$. If so, return x' .
- (2) Else, initialize the (point) variables $h_l := (-\infty, 0)$ and $h_r := (+\infty, 0)$.
- (3) Now for all $G \in \mathcal{S} \setminus \{H\}$, do:
 - (3.1) Let y be the intersection point of the lines ℓ_G and ℓ_H
 - (3.2) If G points right and y has larger x -coordinate than h_l , update $h_l := y$
 - (3.3) If G points left and y has smaller x -coordinate than h_r , update $h_r := y$
- (4) If h_l has strictly larger x -coordinate than h_r , return 'empty', else return h_l .

That the running time is linear is trivial.

Now we argue for correctness. Suppose $R(\mathcal{S}) \neq \emptyset$. Since the conditions of (a) are then satisfied, we know that $c(\mathcal{S})$ is either equal to x' or on ℓ_H . In the first case, the algorithm returns on line (1). In the second case, the algorithm maintains two boundary points h_l and h_r such that on ℓ_H , exactly the points between h_l and h_r are admissible. Since we know $c(\mathcal{S}) \in \ell_H$ and $c(\mathcal{S}) \in R(\mathcal{S})$, we must have $c(\mathcal{S}) = h_l$ because this is the point of smallest x -coordinate within that set.

On the other hand if $R(\mathcal{S}) = \emptyset$, then we *must* end up in the situation that h_1 lies right of h_r as otherwise h_1 , for instance, would be contained in all halfspaces contrary to the assumption.

(c) Consider the following algorithm.

Solve. Input: \mathcal{S}

- (1) If $\mathcal{S} = \{H_1, H_2\}$, return g as the certificate.
- (2) Else, select $H \in \mathcal{S} \setminus \{H_1, H_2\}$ uniformly at random.
- (3) Recursively call **Solve**($\mathcal{S} \setminus \{H\}$).
- (4) If the result is 'empty', return 'empty'
- (5) Else, if the result is a point x' , call **SolveInc**(\mathcal{S}, H, x') and return the result.

The base case is clearly fine. General correctness then follows by induction from the correctness of **SolveInc**.

The more difficult part will be to determine the running time. At first sight, it looks like this algorithm might require quadratic time; after all, we are calling **SolveInc** linearly many times and we know that **SolveInc** uses up to linear time in the worst case. The crucial observation is that **SolveInc**, on input $(\mathcal{S}, H, c(\mathcal{S} \setminus \{H\}))$, requires linear time only if $c(\mathcal{S}) \neq c(\mathcal{S} \setminus \{H\})$. If on the other hand those certificates are equal, then **SolveInc** can return on line (1) and takes only constant time to terminate.

Let us first only consider sets \mathcal{S} such that $R(\mathcal{S}) \neq \emptyset$. Let t_n be expected running time for $n = |\mathcal{S}|$ such halfplanes. We have $t_2 = \mathcal{O}(1)$. Now for $n \geq 3$, t_n is bounded according to the above observation as

$$t_n \leq t_{n-1} + \Pr[c(\mathcal{S}) \neq c(\mathcal{S} \setminus \{H\})] \cdot \mathcal{O}(n) + \Pr[c(\mathcal{S}) = c(\mathcal{S} \setminus \{H\})] \cdot \mathcal{O}(1).$$

But what is the probability $\Pr[c(\mathcal{S}) \neq c(\mathcal{S} \setminus \{H\})]$? According to (a), this can happen only if $\ell_H \ni c(\mathcal{S})$. But since H is chosen uniformly at random from all halfplanes (except for H_1, H_2) and since $c(\mathcal{S})$ cannot lie on any more than two boundary lines, the probability that this happens is bounded by $\frac{2}{n-2}$. We therefore obtain that

$$t_n \leq t_{n-1} + \frac{2}{n-2} \cdot \mathcal{O}(n) + \mathcal{O}(1) = t_{n-1} + \mathcal{O}(1).$$

From this it readily follows that $t_n = \mathcal{O}(n)$ as desired.

The only thing we have left to do is to consider the case when the result will be 'empty'. However, consider that as soon as a recursive step returns 'empty', all higher recursion levels return in constant time. So the transition from a certificate point to the result 'empty' happens at most once in the course of the algorithm. Before that, the bound we just derived applies. Therefore the total time is no bigger than linear in that case either.

Solution 3

Given a linear program in standard form¹,

$$\text{maximize } c^T x \text{ subject to } Ax \leq b, \quad (\text{LP 1})$$

where $A \in \mathbf{R}^{m \times n}$, $c \in \mathbf{R}^n$ and $b \in \mathbf{R}^m$, we can convert it into equational form as follows: First we replace the " \leq " by a " $=$ " by the following trick.

$$\text{maximize } c^T x \text{ subject to } Ax + \varepsilon = b, \quad \varepsilon \geq 0 \quad (\text{LP 2})$$

where $\varepsilon = (\varepsilon_1, \dots, \varepsilon_m)$ is a vector of m new variables. In a second step we get rid of unconstrained (= possibly negative) variables. To this end we replace each x_i by $x'_i - x''_i$, where x'_i, x''_i are two new nonnegative variables:

$$\text{maximize } c^T(x' - x'') \text{ subject to } A(x' - x'') + \varepsilon = b, \quad x' \geq 0, \quad x'' \geq 0, \quad \varepsilon \geq 0. \quad (\text{LP 3})$$

Now we write this LP in such a way that it is undoubtedly in equational form, e.g. like this:

$$\text{maximize } \begin{pmatrix} c \\ -c \\ 0 \end{pmatrix}^T \begin{pmatrix} x' \\ x'' \\ \varepsilon \end{pmatrix} \text{ subject to } \begin{pmatrix} A & -A & I_m \end{pmatrix} \begin{pmatrix} x' \\ x'' \\ \varepsilon \end{pmatrix} = b, \quad \begin{pmatrix} x' \\ x'' \\ \varepsilon \end{pmatrix} \geq 0. \quad (\text{LP 4})$$

In what sense have we "converted" the LP into equational form? We make the following notes.

- If x is a feasible solution of (LP 1), then a corresponding feasible solution of (LP 4) is given by

$$\begin{aligned} x' &= (\max\{x_1, 0\}, \dots, \max\{x_n, 0\}), \\ x'' &= -(\min\{x_1, 0\}, \dots, \min\{x_n, 0\}), \\ \varepsilon &= b - Ax. \end{aligned}$$

Furthermore this feasible solution to (LP 4) has the same objective value as x .

- Correspondingly, if (x', x'', ε) is a feasible solution of (LP 4), then $x' - x''$ is a feasible solution of (LP 1) with the same objective value.

Complexity: The linear program (LP 4) has $2n + m$ variables and $2n + 2m$ constraints, where the original (LP 1) had n variables and m constraints.

¹You might want to revise at this point how to convert any linear program into standard form (section 6.1 of the lecture notes).